


## Dynamic Traffic Assignment Basics Definitions and Properties

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**TeQson Lab** | Transportation Equilibrium  
Simulation and  
Optimized Networks

## Motivation

- DTA is emerging as a practical tool for numerous planning and operational applications
- Due to the recent advances in DTA, uncertainty remains on model capabilities, implications, etc
- The Primer attempts to address the core issues in DTA models



## Motivation: What is a “good” DTA?

- A good model must provide sufficiently sound predictions of system behavior given the necessary resource constraints
- DTA provides a superior representation of traffic, dynamic conditions, and resulting behavior.
  - These benefits come at a computational and implementation cost



## Motivation: What is a “good” DTA?

- Are the implementation costs worth the additional descriptive benefit?
- To answer, we must understand the underlying properties
- To understand a DTA approach fully the following must be precisely defined:
  - formulation assumption
  - solution method and
  - traffic model/simulation



## Defining DTA

- Three things must be defined fully
  - Problem formulation
  - Solution method
  - Traffic model
- Why is there lack of consensus?
  - “Relatively” new
    - Well over a decade now of work, but there are lags between academic research and implementation.
  - Disconnect between research and practice
  - Solution methods more complex for dynamic than static
- **Somewhat similar evolution for static decades ago**



## Defining DTA

- Static progressed from “primitive” approaches to more advanced solution methods (Sheffi, 1985) over the course of many years:
  - All-or-nothing
  - Capacity restrained
  - Modified capacity restrained
  - Incremental Assignment
  - MSA
  - **Frank-Wolfe**
  - Many newer advances past FW



## Static to DTA: What should NOT Change

- **Formulation**
- There are two primary assumptions of equilibrium (whether static or dynamic)
  - Users are “greedy”
  - Users are familiar with the system
- While special cases and extensions may be considered (familiarity might be tuned with SUE for instance)
- These assumption should not change fundamentally simply because time is being considered



## Formulation/Assumptions

- Behavioral assumptions are critical
- Equilibrium represents one of the ***simplest*** cases of behavior
  - *But equilibrium requires iteration*
- Without sound behavioral assumptions **transferability** and **consistency** are not achievable
  - Calibration alone (without sound behavior) does **not** imply transferability and consistency



## Static to DTA: What should NOT Change

- **Solution Method**
- Static methods often employ ***at least*** Frank-Wolfe
- While perhaps not ideal, they provide some measure of
  - Convergence criteria
  - Efficiency
  - Consistency
- These can not be lost in an attempt to model temporal behavior



## Why worry about formulation/method?

- Based on the assumptions of the problem formulation and the solution method
  - Substantially different results will be observed
  - Consistency will also alter dramatically and may be fully unachievable



## Examples for Static

- For DTA, we first need to be sure on static methods
- First we need a static formulation
- This has been well established

$$\min \sum_a \int_0^{x_a} c_a(\omega) d\omega$$

s.t.

$$\sum_k h_k^{rs} = q_{rs} \quad \forall r, s$$

$$h_k^{rs} \geq 0 \quad \forall k, r, s$$

$$x_a = \sum_r \sum_s \sum_k h_k^{rs} \delta_{a,k}^{rs} \quad \forall a$$



## For Static: Solution Methods

- Many solution methods work by “linearizing” the non-linear objective.
- The linearized version is simply the gradient of the function
- For UE, we are very lucky.
- Basically, by taking derivatives, the integrals disappear, leaving simply the cost functions!
- Therefore, to solve the linear problem we simply need to find the shortest path!
- This must be done many times though.



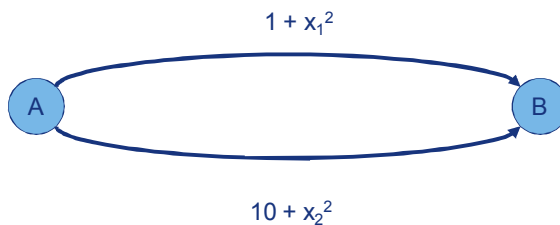
## Static: Example of solution methods

- Quick examples of
  - Method of Successive Averages (MSA)
  - Frank Wolfe (FW)
- Closely see differences from solution method
  - Even when assumptions/formulation the same
- For DTA analogy



## MSA

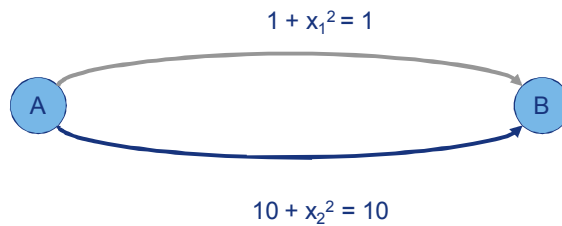
Demand A,B = 10



Begin by taking free-flow costs



## MSA – Iteration 1



So All-or-nothing flow assignment is

$$x_1' = 10 \text{ and } x_2' = 0$$

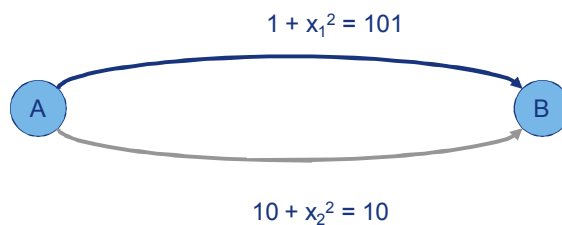
$$\text{or } x = [10 \ 0]$$

Calculate new costs based on these flows



## MSA - Iteration 2

$$x = [10 \ 0]$$



So All-or-nothing flow assignment is  $x_1' = 0$  and  $x_2' = 10$

$$\text{or } x_i = \left(\frac{1}{2}\right) x' + \left(1 - \frac{1}{2}\right) x_{\text{old}} \quad \text{so}$$

$$x_1 = \left(\frac{1}{2}\right) 0 + \left(\frac{1}{2}\right) 10 = 5$$

$$x_2 = \left(\frac{1}{2}\right) 10 + \left(\frac{1}{2}\right) 0 = 5$$

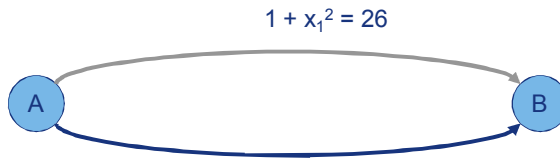
Calculate new costs based on these flows





## MSA - Iteration 3

$$x = [5 \ 5]$$



$$10 + x_2^2 = 35$$

So All-or-nothing flow assignment is  $x_1' = 10$  and  $x_2' = 0$

or  $x_i = (1/3) x' + (1 - 1/3) x_{old}$  so

$$x_1 = (1/3) 10 + (2/3) 5 = 6.66666$$

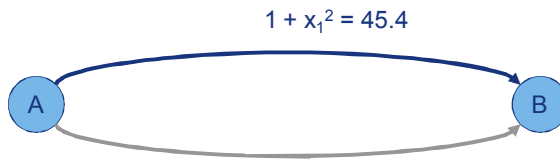
$$x_2 = (1/3) 0 + (2/3) 5 = 3.33333$$

Calculate new costs based on these flows



## MSA - Iteration 4

$$x = [6.666666 \ 3.333333]$$



$$10 + x_2^2 = 21.9$$

So All-or-nothing flow assignment is  $x_1' = 0$  and  $x_2' = 10$

or  $x_i = (1/4) x' + (1 - 1/4) x_{old}$  so

$$x_1 = (1/4) 0 + (3/4) 6.66666 = 5$$

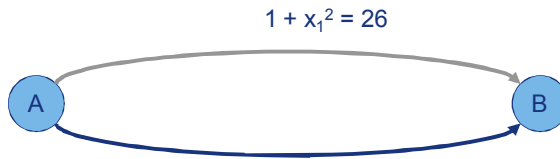
$$x_2 = (1/4) 10 + (3/4) 3.33333 = 5$$

Calculate new costs based on these flows



## MSA - Iteration 5

$$x = [5 \ 5]$$



$$10 + x_2^2 = 35$$

So All-or-nothing flow assignment is  $x_1' = 10$  and  $x_2' = 0$

or  $x_i = (1/5) x' + (1 - 1/5) x_{old}$  so

$$x_1 = (1/5) 10 + (4/5) 5 = 6$$

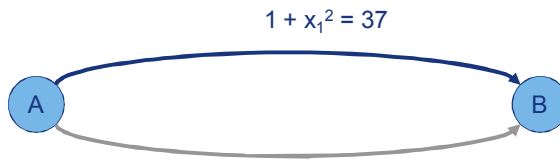
$$x_2 = (1/5) 0 + (4/5) 5 = 4$$

Calculate new costs based on these flows



## MSA - Iteration 6

$$x = [6 \ 4]$$



$$10 + x_2^2 = 26$$

So All-or-nothing flow assignment is  $x_1' = 0$  and  $x_2' = 10$

or  $x_i = (1/6) x' + (1 - 1/6) x_{old}$  so

$$x_1 = (1/6) 0 + (5/6) 6 = 5$$

$$x_2 = (1/6) 10 + (5/6) 4 = 3.33333$$

Calculate new costs based on these flows



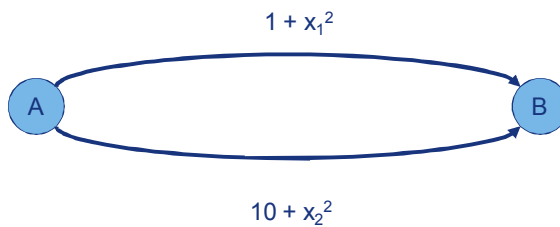
## MSA

- Iter 7            Costs = [ 26    35 ]
- Iter 8            Costs = [ 37    26 ]
- Iter 9            Costs = [ 26    35 ]
- Iter 10           Costs = [ 31.9 29.7 ]



## FW

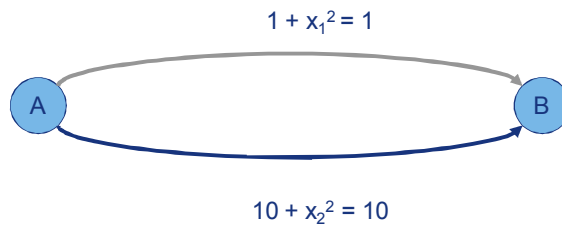
Demand A,B = 10



Begin by taking free-flow costs



## FW – Iteration 1 – the same



So All-or-nothing flow assignment is

$$x_1' = 10 \text{ and } x_2' = 0$$

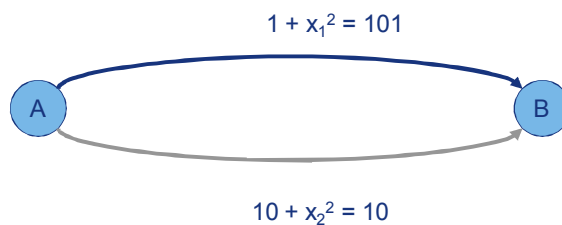
$$\text{or } x = [10 \ 0]$$

Calculate new costs based on these flows



## FW - Iteration 2

$$x = [10 \ 0]$$



So All-or-nothing flow assignment is  $x_1' = 0$  and  $x_2' = 10$

$$\text{or } x_i = (\lambda) x' + (1 - \lambda) x_{\text{old}} \quad (\text{the difference})$$

But how to find  $\lambda$ ?

Calculate new costs based on these flows



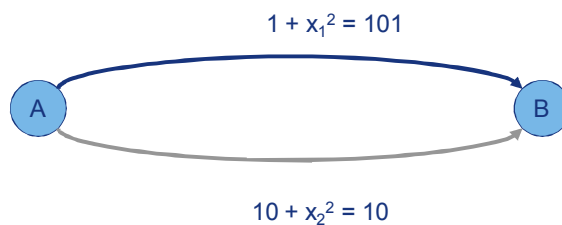
## Finding $\lambda$

- Regardless of network size, number of zones, etc. there is a **single** variable  $\lambda$
- Finding a single variable is relatively easy as long as we have an **objective  $f(\lambda)$**
- For static that comes directly from the formulation we stated



## FW - Iteration 2

$$x = [10 \ 0]$$



So All-or-nothing flow assignment is  $x_1' = 0$  and  $x_2' = 10$

or  $x_i = (\lambda) x' + (1 - \lambda) x_{old}$  (the difference)

$\lambda$  can be found = .455

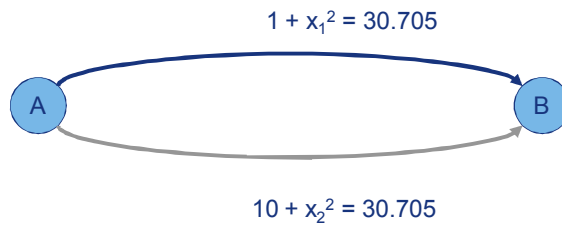
So  $x = [5.45 \ 4.55]$

Calculate new costs based on these flows



## FW - Iteration 3

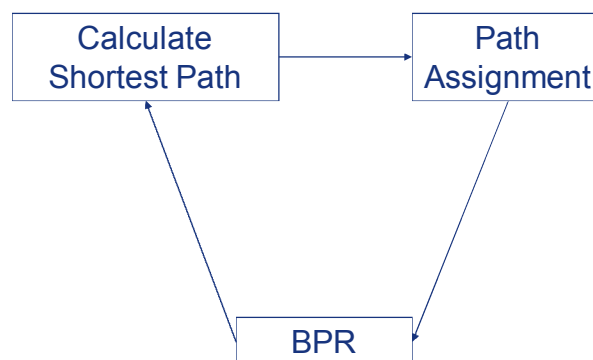
$x = [5.45 \ 4.55]$



Stop



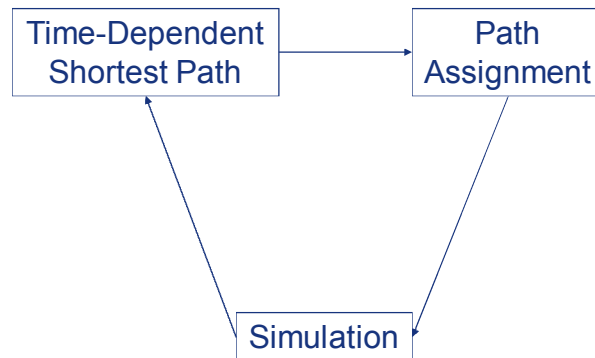
## MSA/FW Static Assignment



So, how do we move to DTA?

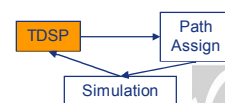


## Simulation-based DTA



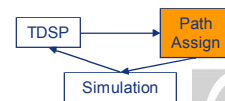
### Simulation-based DTA: Time-Dependent Shortest Path

- Analogous to Shortest Path in static UE
  - Given an arrival time at destination (or departure time from origin), find shortest path
  - Link travel time depends on time of arrival
  - Relatively high fidelity may be needed to capture the impact of control, etc



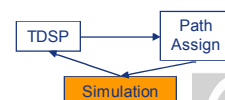
## Simulation-based DTA: Path Assignment Component

- Multiple emerging DTA methods include:
- Method of Successive Averages
  - Uses fixed path splits over iterations
- Simplicial Decomposition
  - Employs objective functions
- Other “gap-based” methods



## Supply: Desired Features of a Traffic Flow Model

- Ability to model:
  - Bottlenecks
  - Link Spillover
  - Shockwave Propagation
- The correct modeling of these issues are critical to Dynamic Traffic Assignment

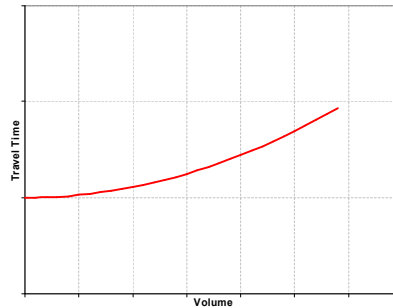




## Modeling Congestion

A typical BPR volume-delay function:

$$t(v) = t_o \cdot \left( 1 + \left( \frac{v}{c} \right)^\alpha \right)$$



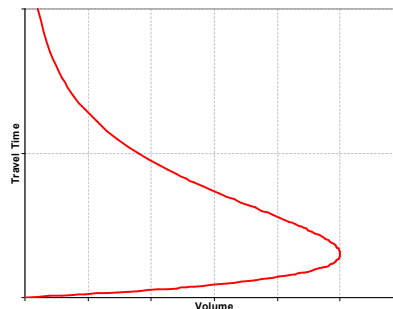
## Modeling Congestion

Even simple traffic models provide a more accurate representation.

$$u(k) = u_f \cdot \left( 1 - \frac{k}{k_j} \right)$$

$$q = u \cdot k$$

$$t(u) = \frac{t_o \cdot u_f}{u}$$

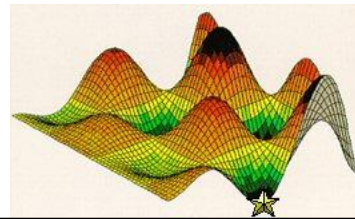


Given the same flow, two travel times are possible.  
The difference is **density**.



## Convergence

- Ideally, we would like to know:
  - “How far is the current solution from equilibrium?”
- Running “until things stop changing” may be problematic
  - Convergence of system cost may not imply convergence of link flow
  - Convergence of link flow may not imply equilibrium path costs



## Simulation-based DTA: Path Assignment: Distance to Equilibrium

- DTA convergence offers opportunities
- An example of an equilibrium “gap” function
- Given an O-D with three used routes A, B and C:

Path	Flow	Travel Time	Least Travel Time	Gap
A	10	5	3	20
B	20	6	3	60
C	30	3	3	0

80

## Research Comparing DTA to Static

- It is surprisingly difficult to compare static and dynamic
  - If attempted incorrectly will greatly confound consistency of analysis
  - Another impact of  $V/C > 1$  issues and such
- Also an issue because of boundary conditions/transient behavior
- If peak period analysis is attempted with DTA, a warm-up cool-down period is required
  - 24-hr modeling may be best approach




## Sampling of Other Models

- The previous approach to DTA is essentially:
- **Equilibrium on Experienced Travel Cost**
- Numerous other models exist
  - Non-equilibrium approaches
  - “One-shot” Models
  - Instantaneous Travel time Models



## Instantaneous Travel Time Versus Experienced Travel Time Modeling



Time	Link Travel Times		
0	1	2	3
1	2	3	4
2	1	4	4
3	1	3	5
4	2	3	5

- If departing at time 1:
- **Instantaneous travel time = 6** (just add up first row)
- Experienced travel time =
  - Link 1 travel time (at time 0) = 1
  - Link 2 travel time (at time 1) = 3
  - Link 3 travel time (at time 4) = 5
- **Experienced Travel Time = 9**



## One-shot Modeling

- “One-shot” models do not attempt an equilibrium in the previously described sense
- Traffic does spread over routes but not due to equilibration
  - One approach is to base route choice decisions at time T no congestion up to time T
  - For instance, use instantaneous travel times
- Most similar to incremental approaches for static assignment



### Harsh Truth #1: Disequilibrium Versus Non-Convergence

- It is occasionally noted that traffic isn't really in equilibrium so why worry about it?
- However, if we want to use DTA for planning then we must have stable solutions
  - If noise or randomness substantially impact the solution, the results are not defensible in a planning context
  - Further, explainable behavior is important



### Harsh Truth #1: Disequilibrium Versus Non-Convergence

- There is ongoing research into “disequilibrium” (or transient) traffic modeling.
  - However, this research is clearly distinct from **non-convergence**
  - Put simply, stopping an equilibrium model prior to convergence is not correct (and is not supported by arguments related to traffic disequilibria)
  - For planning applications, any disequilibrium network model must still provide a stable solution (and current options for this appear limited)
  - Equilibrium remains the simplest approach to generate stable solutions for planning applications



## Harsh Truth #2: Costs of DTA

- If a DTA approach does not result in substantially higher computational costs, there are two highly likely reasons:
  - Very few (if any) behavioral advantages are being obtained
  - Substantial sacrifices are being made in terms of solution quality, convergence, stability, etc
- However, many questions can only be answered given the superior behavioral representation of DTA



## Conclusions

- A wide variety of models may currently be termed DTA
- Without any modifying term, the Primer defines DTA as an **Equilibrium Based on Experienced Travel Cost**
- Equilibrium remains the simplest approach to obtain stable solutions for planning applications
- Convergence and stability are still absolutely critical for planning applications (using DTA should not change this)
- As a new consideration for planners, traffic realism is also critical





THANK YOU!

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